



## Sensitivity Analysis: A Variational Approach

François-Xavier Le Dimet, Innocent Souopgui, M.Yussuf Hussaini, Ha Tran  
Thu

### ► To cite this version:

François-Xavier Le Dimet, Innocent Souopgui, M.Yussuf Hussaini, Ha Tran Thu. Sensitivity Analysis: A Variational Approach. 25th Biennial Conference on Numerical Analysis, Jun 2013, Glasgow, United Kingdom. 2013. hal-00932570

**HAL Id: hal-00932570**

**<https://inria.hal.science/hal-00932570>**

Submitted on 20 Jan 2014

**HAL** is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers.

L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

# Sensitivity Analysis : A Variational Approach

F.-X. Le Dimet(1,2), I. Souopgui(2), Tran Thu Ha(3), M. Y.  
Hussaini(2)

(1) Université de Grenoble

(2) Florida State University

(3) Institute of Mechanics, Vietnamese Academy of Sciences, Hanoi

*ledimet@imag.fr*

September 30, 2013

- General Sensitivity Analysis
- Sensitivity and Data Assimilation
- Second Order Analysis
- A 1-D Example
- An Application to a pollution problem

# Sensitivity Analysis: Deterministic Approach

- Model:  $\mathcal{F}$ :

$$\mathcal{F}(\mathcal{X}, \mathcal{U}) = 0 \quad (1)$$

- Scalar Response Function  $\mathcal{G}$ :

$$\mathcal{G}(\mathcal{X}, \mathcal{U}) \quad (2)$$

- Sensitivity  $\mathcal{S}$  is by definition the gradient of  $\mathcal{G}$  with respect to  $\mathcal{U}$ :

$$\mathcal{S} = \nabla \mathcal{G}(\mathcal{X}(\mathcal{U}), \mathcal{U}) \quad (3)$$

- An adjoint variable  $\mathcal{P}$  is introduced as the solution of :

$$\left[ \frac{\partial \mathcal{F}}{\partial \mathcal{X}} \right]^t . \mathcal{P} = \left[ \frac{\partial \mathcal{G}}{\partial \mathcal{X}} \right] \quad (4)$$

- Then we get :

$$\mathcal{S} = \left[ \frac{\partial \mathcal{G}}{\partial \mathcal{U}} \right] - \left[ \frac{\partial \mathcal{F}}{\partial \mathcal{U}} \right]^t . \mathcal{P} \quad (5)$$

# Data Assimilation for Pollution Modeling

- $X$  is the state variable (velocity, surface elevation) governed by :

$$\begin{cases} \frac{dX}{dt} = F(X) \\ X(0) = U \end{cases} \quad (6)$$

- The concentration of pollutant  $C$ , produced by sources  $S$  verifies:

$$\begin{cases} \frac{dC}{dt} = G(X, C, S) \\ C(0) = V \end{cases} \quad (7)$$

- $U$  and  $V$  are unknown. The VDA problem is to evaluate them from observation  $X_{obs}$  and  $C_{obs}$ , in order to minimize the cost function  $J$  defined by:

$$J(U, V) = \frac{1}{2} \int_0^T \|EX - X_{obs}\|^2 dt + \frac{1}{2} \int_0^T \|DC - C_{obs}\|^2 dt \quad (8)$$

- For sake of simplicity regularization terms, of great practical importance, are not displayed

# Data Assimilation for Pollution Modeling: Optimality System

- $P$  and  $Q$  adjoint variables are introduced as the solution of the system :

$$\begin{cases} \frac{dP}{dt} + \left[ \frac{\partial F}{\partial X} \right]^t . P + \left[ \frac{\partial G}{\partial X} \right]^t . Q = E^t (EX - X_{obs}) \\ P(T) = 0; \end{cases} \quad (9)$$

;

$$\begin{cases} \frac{dQ}{dt} + \left[ \frac{\partial G}{\partial C} \right]^t . Q = D^t (DC - C_{obs}); \\ Q(T) = 0, \end{cases} \quad (10)$$

- Then the gradient of  $J$  with respect to  $U$  and  $V$  are given by :

$$\nabla J_U = -P(0) \quad (11)$$

$$\nabla J_V = -Q(0) \quad (12)$$

# Sensitivity with respect to Observations and Sources

- If some response function  $\mathcal{S}$  is introduced, how to evaluate the sensitivity with respect to observations? For instance how to evaluate the impact of an error of observation on a prediction?
- What should be the "model"  $\mathcal{F}$  of the general sensitivity analysis?
- Because only the Optimality System contains the observation, the sensitivity analysis must be carried out on the O.S. considered as a Generalized Model
- Deriving the O.S. leads to carry out a **Second Order Analysis**.



# Computing the sensitivity with respect to sources : second order adjoint.

- We need to introduce four second order adjoint variables  $\Gamma$ ,  $\Lambda$ ,  $\Phi$  and  $\Psi$  as the solution of :

$$\left\{ \begin{array}{l} \frac{d\Gamma}{dt} + \left[ \frac{\partial F}{\partial X} \right]^t \cdot \Gamma + \left[ \frac{\partial F}{\partial X} \right]^t \cdot \Lambda + \left[ \frac{\partial^2 F}{\partial X^2} P \right]^t \cdot \Phi \\ \quad + \left[ \frac{\partial^2 G}{\partial X^2} Q \right]^t \cdot \Phi + \left[ \frac{\partial^2 G}{\partial C \partial X} Q \right]^t \cdot \Psi - E^t E \Phi = 0; \\ \Gamma(0) = 0; \\ \Gamma(T) = 0, \end{array} \right. \quad (13)$$

# Computing the sensitivity with respect to sources 2



$$\left\{ \begin{array}{l} \frac{d\Lambda}{dt} + \left[ \frac{\partial F}{\partial C} \right]^t \cdot \Lambda + \left[ \frac{\partial^2 G}{\partial C \partial X} Q \right]^t \cdot \Phi \\ \quad + \left[ \frac{\partial^2 G}{\partial X^2} Q \right]^t \cdot \Psi - D^t D\Psi = \frac{\partial \varphi}{\partial C}; \\ \Lambda(0) = 0; \\ \Lambda(T) = 0, \end{array} \right. \quad (14)$$

$$\frac{d\Phi}{dt} + \left[ \frac{\partial F}{\partial X} \right]^t \cdot \Phi = 0, \quad (15)$$

$$\frac{d\Psi}{dt} + \left[ \frac{\partial G}{\partial C} \right]^t \cdot \Psi = 0, \quad (16)$$

• Then it comes :

$$\nabla \varphi = \left[ \frac{\partial F}{\partial S} \right]^t \cdot \Lambda + \left[ \frac{\partial^2 G}{\partial X^2} Q \right]^t \cdot \Phi + \left[ \frac{\partial^2 G}{\partial C \partial S} Q \right]^t \cdot \Psi + \frac{\partial \varphi}{\partial S} \quad (17)$$

- The sensitivity is obtained by solving the coupled system of four equations
- The System involves second order terms.
- We found a **non-standard problem** : **two equations have two conditions an initial condition and a final condition, the other two equations have no condition**

# Solving the Non-Standard problem

- The Non-Standard problem can be symbolically written :

$$\begin{cases} \frac{dX}{dt} = K(X, Y), & t \in [0, T]; \\ \frac{dY}{dt} = L(X, Y), & t \in [0, T] \end{cases} \quad (18)$$

- with :

$$\begin{cases} X(0) = 0; \\ X(T) = 0 \end{cases} \quad (19)$$

and no condition on  $Y$ .

NSP is transformed into a problem of optimal control by introducing the control  $U$  and a cost-function  $J_P(U)$  with :

$$\begin{cases} X(0) = 0; \\ Y(0) = U. \end{cases} \quad (20)$$

## Solving the Non-Standard problem 2

A cost function  $J_P(U)$  is defined by:

$$J_P(U) = \frac{1}{2} \|X(T, U)\|^2 + \frac{1}{2} \|U\|^2 \quad (21)$$

If  $Z$  and  $W$  are defined as the solution of:

$$\frac{dW}{dt} + \left[ \frac{\partial K}{\partial X} \right]^t \cdot W + \left[ \frac{\partial L}{\partial X} \right]^t \cdot Z = 0; \quad (22)$$

$$\frac{dZ}{dt} + \left[ \frac{\partial K}{\partial Y} \right]^t \cdot W + \left[ \frac{\partial L}{\partial Y} \right]^t \cdot Z = 0; \quad (23)$$

$$Z(T) = 0; W(T) = X(T), \quad (24)$$

then we get

$$\nabla J_P(U) = -Z(0) + U \quad (25)$$

# Solving the Non-Standard problem 3

This problem involved third derivatives of the original model.  
Recent developments on the NSP have been recently carried out by V. Shutyaev and F.-X. Le Dimet  
The existence of a solution is demonstrated  
Another method to solve NSP is proposed.

# A 1-D Example 1

Let us assume that the one dimensional velocity field  $u = u(x, t)$  evolves according to the Burgers equation given by :

$$\left\{ \begin{array}{l} \frac{\partial u}{\partial t} + u \frac{\partial u}{\partial x} - \nu \frac{\partial^2 u}{\partial x^2} = f, x \in \Omega = ]-1, 1[, t \in [0, T]; \\ u = u_0, t = 0, \\ u = u_1, x \in \{-1, 1\}, \end{array} \right. \quad (26)$$

Evolution of the pollutant's concentration:

$$\left\{ \begin{array}{l} \frac{\partial c}{\partial t} + u \frac{\partial c}{\partial x} = \eta \frac{\partial^2 c}{\partial x^2} + s, x \in ]-1, 1[, t \in [0, T] \\ c = c_0, t = 0; \\ c = c_1, x \in \{-1, 1\} \end{array} \right. \quad (27)$$

# A 1-D Example: cost function

The cost function takes the form (with continuous observation in space and time):

$$J(u_0, c_0) = \frac{1}{2} \int_0^T \|u - u_{obs}\|_{\Omega}^2 dt + \frac{1}{2} \int_0^T \|c - c_{obs}\|_{\Omega}^2 dt. \quad (28)$$

where  $\|f\|_{\Omega}^2 = \int_{\Omega} f(x)f(x)dx = \int_0^1 f(x)f(x)dx$ .



# A 1-D Example: adjoint model

The adjoint variables  $p$  and  $q$  are introduced as the solutions of :

$$\left\{ \begin{array}{l} \frac{\partial p}{\partial t} + u \frac{\partial p}{\partial x} + \nu \frac{\partial^2 p}{\partial x^2} + q \frac{\partial c}{\partial x} = u - u_{obs} \\ p(t = T) = 0 \\ p = 0, x \in \{-1, 1\} \end{array} \right. \quad (29)$$

$$\left\{ \begin{array}{l} \frac{\partial q}{\partial t} + \frac{\partial u q}{\partial x} + \eta \frac{\partial^2 q}{\partial x^2} = c - c_{obs} \\ q(t = T) = 0 \\ q = 0, x \in \{-1, 1\} \end{array} \right. \quad (30)$$

And the gradient of the cost function is given by:

$$\begin{aligned} \nabla_{u_0} J &= -p(0) \\ \nabla_{c_0} J &= -q(0) \end{aligned}$$

# Sensitivity of a response function

Let  $\varphi$  be a function of the concentration and the source functions, the response function is given by:

$$\Phi_A(t, s) = \int_{\Omega_A} \varphi(c, s) dx \quad (31)$$

where  $\Omega_A \subset \Omega$  is the response region. Following the guidelines of the derivation of the gradient, we introduce the adjoint variables  $\Gamma$ ,  $\phi, \psi$  and  $\Lambda$  as the solution of:

# Sensitivity of a response function

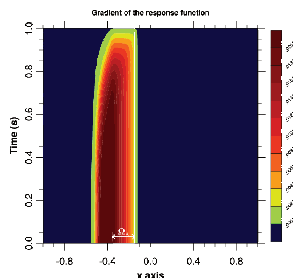
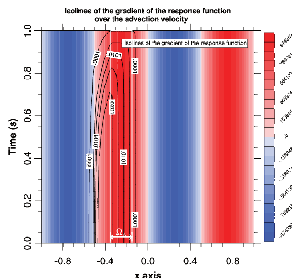
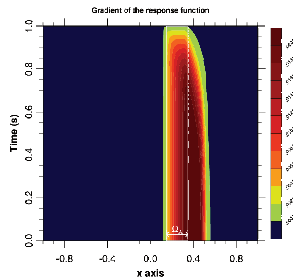
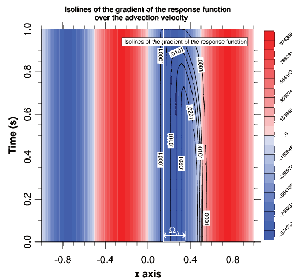
$$\left\{ \begin{array}{l} \frac{\partial \Gamma}{\partial t} + u \frac{\partial \Gamma}{\partial x} + \nu \frac{\partial^2 \Gamma}{\partial x^2} - \Lambda \frac{\partial c}{\partial x} \\ \quad - \phi \frac{\partial p}{\partial x} + q \frac{\partial \psi}{\partial x} - \phi = 0 \\ \Gamma = 0, t \in \{0, T\} \\ \Gamma = 0, x \in \{-1, 1\} \end{array} \right. \quad (32)$$

$$\left\{ \begin{array}{l} \frac{\partial \Lambda}{\partial t} + \frac{\partial u \Lambda}{\partial x} + \eta \frac{\partial^2 \Lambda}{\partial x^2} + \frac{\partial q \phi}{\partial x} - \psi = -\frac{\partial \varphi}{\partial c} \\ \Lambda = 0, t \in \{0, T\} \\ \Lambda = 0, x \in \{-1, 1\} \end{array} \right. \quad (33)$$

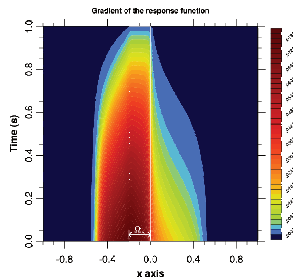
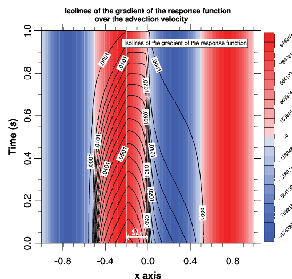
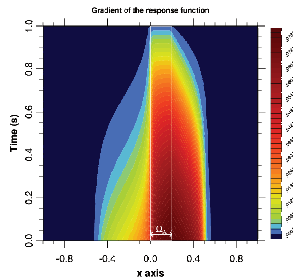
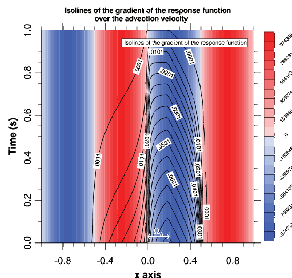
Where the function  $1_{\Omega_A}$  is:

$$1_{\Omega_A}(x) = \begin{cases} 1, & \text{if } x \in \Omega_A \\ 0, & \text{if not.} \end{cases}$$

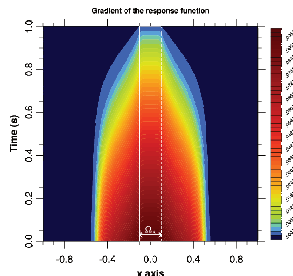
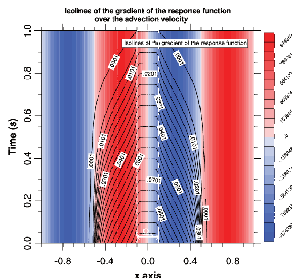
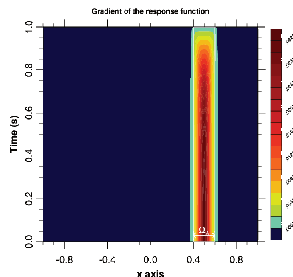
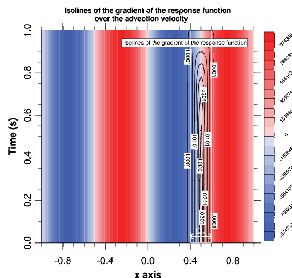
# Numerical results



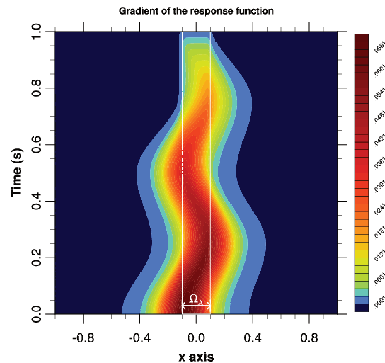
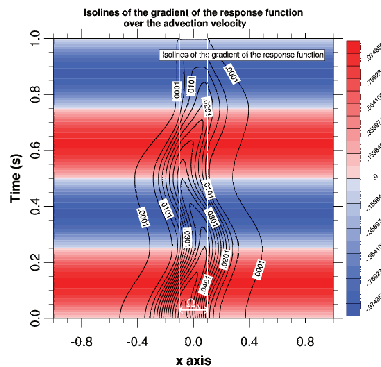
# Numerical results



# Numerical results



# Numerical results



# The End